THE EFFECT OF JPEG COMPRESSION IN CLOSE RANGE PHOTOGRAMMETRY

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ABSTRACT: Digital photogrammetry, using digital camera images, is an important low-cost engineering method to produce precise three-dimensional model of either an object or the part of the earth depending on the image quality. Photogrammetry which is cheaper and more practical than the new technologies such as LIDAR, has increased point cloud generation capacity during the past decade with contributions of computer vision. Images of new camera technologies needs huge storage space due to larger image file sizes. Moreover, this enormousness increases image process time during extraction, orientation and dense matching. The Joint Photographic Experts Group (JPEG) is one of the most commonly used methods as lossy compression standard for the storage purposes of the oversized image file. Particularly, image compression at different rates causes image deteriorations during the processing period. Therefore, the compression rates affect accuracy of photogrammetric measurements. In this study, the close range images compressed at the different levels were investigated to define the compression effect on photogrammetric results, such as orientation parameters and 3D point cloud. The outcomes of this study show that lower compression ratios are acceptable in photogrammetric process when moderate accuracy is sufficient.

Keywords: Photogrammetry, JPEG, compression, 3D modeling, image matching, exterior orientation

1. INTRODUCTION

Photogrammetry includes scientific methodologies that calculate the three-dimensional coordinates of an object via measuring corresponding points on the overlapping images. The mathematical relation between an image point and an object point is derived by collinearity equations that based on central projection (Kyle, 2013). The recent integration of computer vision algorithms and photogrammetric methods is leading to interesting procedures which have increasingly automated the entire image-based 3D modelling process (Remondino et al., 2014). In last two decades, Close-Range Photogrammetry (CRP) as a contribution of photogrammetry and computer vision, spread into many fields of engineering applications such as medical modelling applications (Xiao et al., 2014), orthophotos by Unmanned Aerial Systems (UAS) (Akcay, 2015) and documentation of cultural heritages (Yılmaz et al., 2007). Three dimensional textured models, digital surface models and true orthophotos can be produced using advantages of low-cost CRP software.

CRP software implement automatic point detection and robust matching algorithms for photogrammetric process. The most common algorithm; Scale Invariant Feature Transform (SIFT) is a digital property extraction method that allows automatic identification of characteristic points (Lowe, 2004). Bay et al. (2006), developed Speeded Up Robust Features (SURF), that approximate SIFT algorithm with respect to repeatability, distinctiveness, and robustness, yet can be computed and compared much faster. This is achieved by relying on integral images for image convolutions; by building on the strengths of the leading existing detectors and descriptors by using a Hessian matrix-based measure for the detector, and a distribution-based descriptor. But these algorithms might be inadequate when very high accuracy (sub-pixel) is necessary, such as industrial (Luhmann et al., 2015) and engineering photogrammetric measurements (Avsar et al, 2015). Traditional photogrammetric measurement is to mark corresponding points manually, which requires more time-consuming and labour-intensive process. Furthermore, the accuracy of the manual measurement depends on user experience and therefore required product quality may not be achieved. Third measurement method in CPR is the automatic measurement of specially coded targets within sub-pixel accuracy. These approximations depend on combined usage of normalised cross correlation and least square image matching. The size and shape of the specially coded targets are determined according to camera-object distance (Yılmaztürk, 2011).

After image orientation implementation, all three mentioned methods produce three-dimensional model from point cloud by Multi-view stereo (MVS) (Seitz et al., 2006). Multi-view stereo algorithms provide feature detection and are able to construct highly detailed 3D models from multiple images (Furukawa and Ponce, 2007).

Raw data format is the uncompressed or possibly least processed format of the images obtained by digital cameras or scanners. Besides raw formats; camera manufacturers provide lower-file size and high-quality Joint Photographic Experts Group (JPEG) (Hamilton, 1992) and its derivative such as JPEG2000 (Christopouloset al., 2000) images by embedded image processing software in digital camera (Hamilton, 1992). Today, new generation photogrammetric software can process both raw and compressed image files.

Many processing steps applied sequentially while getting JPEG images. First, the colour space conversion is performed. Red, Green, Blue (RGB) image within the visible wavelength color-coded format is converted to $Y'C_bC_r$ (Y' is luma component and C_b, C_r are two chroma components) expressed in the colour space. Colour space conversion, is defined in ITU-R BT.601 standard, previously called CC 601. In second step of the JPEG conversion, especially the colour sub-sampling process is performed to reduce the data flow rate (Kerr, 2012). In the third step, each channel of the image with the block parsing process is divided into 8x8 or 16x16 blocks. The two-dimensional discrete cosine transform (DCT) is applied on each block to determine the energy distribution (Ahmed et al., 1974). Finally, entropy coding with Huffman is calculated for each block (Huffman, 1952) after quantization step is implemented.

Cronk (2001) in his study examined the effect of JPEG compression in CRP. Yılmaztürk and Akcay (2005) examined the JPEG compression effect on the sub-pixel measurement by using different target sizes. Beside photogrammetric studies, Liang et al. (2006) discussed the effect of compression on remote sensing processes. In the study, the effect of different JPEG image compression ratios on the accuracy of the photogrammetric evaluation and generation of three-dimensional point cloud were examined.

2. METHODOLGY AND APPLICATION

2.1 Exterior orientation with different methods

In order to compare the different photogrammetric measurement methods (manual, SURF and coded targets) explained in the introduction; a statue in the university campus was selected as the study object. The raw images were acquired with a 18MP Canon EOS 650D without compression. 37 acquired images were evaluated during the study. The images were oriented according to the mentioned methods (Table 1).

Table 1. Comparison of exterior orientation parameters

Method	Mean square error (m)			# points (avg)
	X	Y	Z	
Manual	0.00090	0.00063	0.00084	29
SURF	0.01041	0.00654	0.00957	680
Coded T.	0.00096	0.00066	0.00087	18

Method	Mean square error (degree)		
	Omega	Phi	Kappa
Manual	0.03295	0.02586	0.03678
SURF	0.05423	0.04119	0.06233
Coded T.	0.03409	0.02654	0.03958

SURF algorithm is able to define characteristic features which have intensive radiometric and spatial changes (Figure 1). As shown in Table 1; approximately 680 points matched in images with SURF algorithm which is 25-35 times more than two other measurement methods. On the contrary; SURF algorithm delivers mean square error of exterior orientation parameters ten times greater than others.

Similar accuracy results were obtained in orientation when coded targets and manual measurement were compared.

Figure 1. SURF feature extraction results.

2.2. Application of Image Compression

The effect of the compression was considered for automatic measurement methods. Therefore, manual measurements were not examined for the compressed images. On the other hand, SURF and coded targets as automatic measurement methods were accounted for different compression levels. Raw data format was converted to JPEG compressed files to obtain ten percentage gradual decreased quality images.

Compression rates especially decreases the amount of the extracted features as well as robust matching number (Chao et al., 2013). However, results showed an unworthy decline in the context of extracted features due to quality loss as shown in Figure 2.
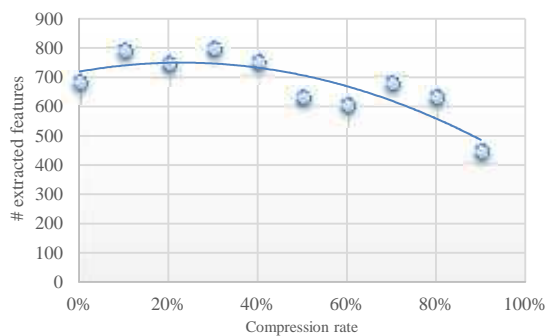

Figure 2. SURF feature extraction results.

Consideration of the extracted features at different compression levels might not be enough in order to

infer the influence of the compression. Understanding of the matching performance is also important beside the feature extraction. Low matching performance is clearly seen at more than forty percentage compression rates when Figure 3 is inspected.
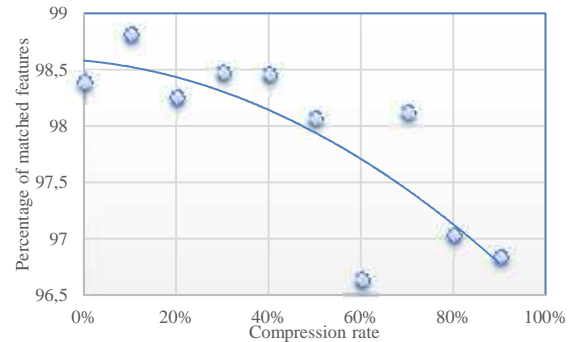

Figure 3. Matched points for SURF.

Coded targets which provide sub-pixel accuracy matching do not give the possibility of an appropriate compression analysis. Automatic photogrammetric measurement with coded target is vulnerable to compression process as the mentioned method is so sensitive to image quality. More than twenty percentage compression makes the photogrammetric calculation failure because of the insufficient matched coded target.

3. DISCUSSION

3.1. Exterior orientation with different methods

In the study, mean square errors (MSE) of the exterior orientation obtained using automatic measurement methods were also compared at each compression levels. Figure 4 illustrates MSE of the projection centre coordinates X_0, Y_0, Z_0 calculated using SURF extraction and matching while Figure 5 indicates MSE of the orientation angles ϖ, φ, κ. As seen in Figure 4, errors of projection centre coordinates sharply increase after forty percent compression level. Figure 5 also indicates that forty percentage compression is determining level during the computation of the rotation angles.
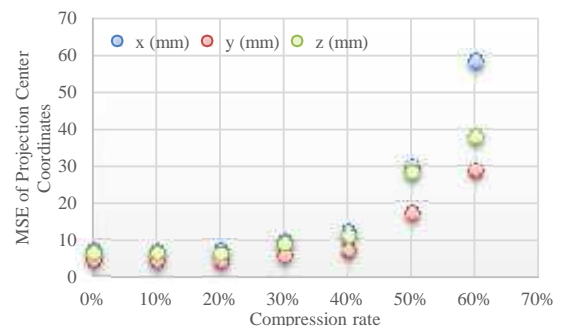

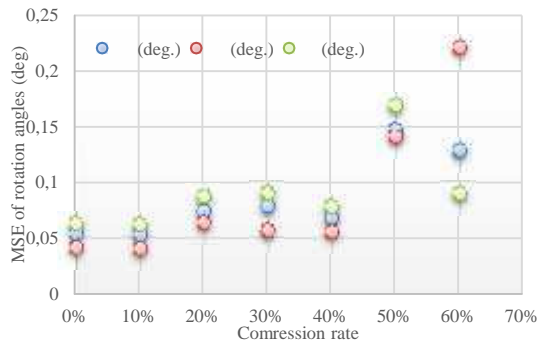Figure 4. SURF feature extraction results.

Figure 5. SURF feature extraction results.

An appropriate compression analysis of exterior orientation with coded targets is as not possible as computed with SURF algorithm. Orientations computed with sub-pixel measurement accuracy showed significant deficiencies at low compression rates. Moreover, high compression rates more than twenty percent were not possible to define orientation values. Figures 6 and 7 illustrate that the quality change in images had little effect on orientation accuracy.
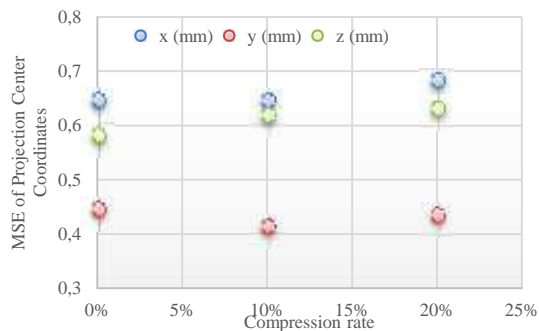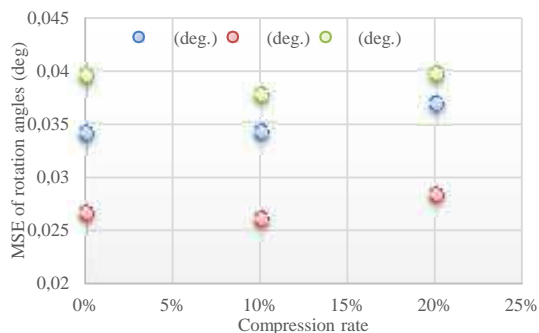
Figure 6. Coded targets feature extraction results.

Figure 7. Coded targets feature extraction results.

3.2. Point cloud generation

Point cloud generation is implemented with MVS also known as dense matching using multiple images. Accuracy of the obtained point is related with the number and angle of image. Figure 8 shows a part of the point cloud from the statue. As it is seen in the figure, point cloud is so dense. Consequently, the photogrammetric point cloud might be a low-cost alternative to LIDAR point cloud in some cases (Nouwakpo et al., 2015).

Figure 8. Point Cloud produced by MVS from SURF results.

Applied compression levels also affect results of MVS point cloud beside exterior orientation. This affect emanates from both faulty exterior orientation results and MVS processing with low resolution images. At the high compression levels, gross errors were observed on the point clouds. Figure 9 shows a point cloud which was obtained in fifty percent compression level. The figure explicitly illustrates the failures on the point cloud. However low compression levels less than fifty percent output more stable point clouds as indicated in Figure 10.

Figure 9. Gross errors of point cloud obtained from %50 compressed images by SURF.

Figure 10. 3D model obtained from %20 compressed images by SURF.

As MVS results were investigated for each compression level, point numbers generated in the clouds were decreased due to higher compression levels. Particularly, numbers of points diminished at the compression higher than thirty percent. On the contrary, the triangulation and filtered point clouds remained still despite compression (Figure 11). Because triangulation and filtered point cloud were computed independently from images, they were not influenced as much as the original point clouds. Gross errors on the MVS point cloud, also reflect the distortions to the textured models.
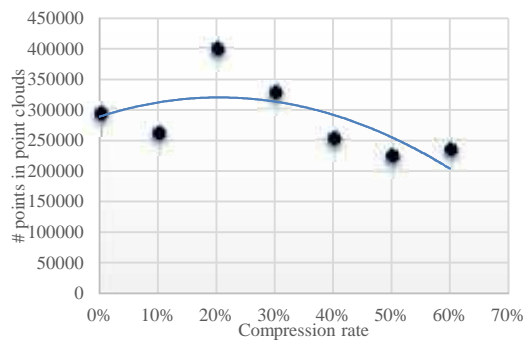

Figure 11. Compression-point graph.

4. RESULTS AND RECOMMENDATIONS

In the experimental tests, it has been shown that the low compression ratios have a negligible effect on the photogrammetric process. On the other hand, if it is considered that a small amount of compression rates can be seriously reduced in file sizes, up to twenty or thirty percent of compression ratio, the photogrammetric modelling been has proved to be problem-free for the works that do not require high precision. However, sub-pixel measurement method showed bad results with low compressed images as the method needed high resolution.

In the case of high compression ratios, the three orientation methods applied also considerably reduced the precision of the exterior orientation and caused significant rough errors in the 3D models. It has been observed that the number of point clouds decreases considerably after thirty percent compression.

No colour subsampling is done in JPEG compression performed in this study. By applying the colour subsample in many different combinations, its effect on the photogrammetry can be revealed in future studies. Also in future work, TIFF format and JPEG compression formats, which are obtained without loss from the raw data format, can be compared. Although JPEG is one of the most used standards, the impact on photogrammetric evaluation of other popular image compression algorithms such as JPEG2000 should be discussed.

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